A sufficient condition for this to hold is that $k \ge k(n, \varepsilon) = (2 + \varepsilon) \frac{\ln n}{\ln 1/q}$. Thus for large *n*, almost no graph $G \in \mathcal{G}(n, p)$ can have a colouring that would assign the same colour to $k(n, \varepsilon)$ or more nodes. Hence, a proper colouring of almost any $G \in \mathcal{G}(n, p)$ requires at least $\frac{n}{k(n, \varepsilon)} = \frac{\ln 1/q}{2+\varepsilon} \cdot \frac{n}{\ln n}$ colours. \Box

Theorem 7.7 Let $p, 0 be constant. Then for a.e. <math>G \in \mathcal{G}(n, p)$:

$$\omega(G) \in \{d, d+1\},\$$

where d = d(n, p) is the largest integer such that

$$\binom{n}{d}p^{\binom{d}{2}} \ge \ln n.$$

(This implies $d = 2\log_{1/p}(n) + O(\log\log n.)$.)

A graph property Q is an isomorphism-closed family of graphs, i.e. if $G \in Q$ (or "G has Q") and $G \approx G'$, then also $G' \in Q$.

A *threshold function* for a graph property Q is a function $t : \mathbb{N} \to \mathbb{R}$ such that

$$\Pr(G \in \mathcal{G} (n, p(n)) \text{ has } Q) \xrightarrow[n \to \infty]{} \begin{cases} 1, \text{ if } p \succ t, \\ 0, \text{ if } p \prec t, \end{cases}$$

where:

$$p \succ t \Leftrightarrow \lim_{n \to \infty} \frac{p(n)}{t(n)} = \infty,$$
$$p \prec t \Leftrightarrow \lim_{n \to \infty} \frac{p(n)}{t(n)} = 0.$$

Further notation:

$$p \sim t \Leftrightarrow \lim_{n \to \infty} \frac{p(n)}{t(n)} = 1,$$
$$p \approx t \Leftrightarrow p(n) = \Theta(t(n)).$$

Denote: $P_n^Q(p) = \Pr(G \in \mathcal{G}(n, p) \text{ has } Q).$

For technical reasons, we will actually use the following slightly stronger definition for a threshold function: t(n) is a threshold function for graph property Q if



Figure 7: $P_n^Q(p)$ for (a) small, (b) intermediate and (c) large *n*.

for any sequence $n_1 < n_2 < ...$ of graph sizes and $p(n_1), p(n_2), ...$ of associated edge probabilities,

$$\lim_{k \to \infty} \frac{p(n_k)}{t(n_k)} = \infty \Rightarrow P^Q_{n_k}(p(n_k)) = 1, \qquad (*)$$
$$\lim_{k \to \infty} \frac{p(n_k)}{t(n_k)} = 0 \Rightarrow P^Q_{n_k}(p(n_k)) = 0. \qquad (**)$$

A graph property is *monotone* if it is preserved under addition of edges, i.e. if G = (V, E) and G' = (V, E') are graphs such that $E \subseteq E'$ and G has Q, then also G' has Q. For monotone Q it is the case that $p_1 \leq p_2 \Rightarrow P_n^Q(p_1) \leq P_n^Q(p_2)$, so the inverse of $P_n^Q(p)$ is well-defined:

$$p_n^Q(\alpha) =$$
 the smallest p such that $P_n^Q(p) \ge \alpha$.

In fact for monotone Q one can show that $P_n^Q(p)$ is a continuous, strictly increasing function of p, so actually $p_n^Q(\alpha) = unique p$ such that $P_n^Q(p) = \alpha$.

Figure 7 illustrates the evolution of the function P_n^Q , and a corresponding threshold function t(n), for a monotone graph property Q from small to large values of n.

Lemma 7.8 A function t(n) is a threshold for monotone graph property Q if and only if $t(n) \approx p_n^Q(\alpha)$ for all $0 < \alpha < 1$.

Proof. Suppose that t(n) is threshold function for Q, but $t(n) \not\approx p_n^Q(\alpha)$ for some $0 < \alpha < 1$. Denoting for brevity $p(n) = p_n^Q(\alpha)$, this means that either there is a sequence n_1, n_2, \ldots such that

$$p(n_k)/t(n_k) \to \infty$$

or there is a sequence n_1, n_2, \ldots such that

$$p(n_k)/t(n_k) \rightarrow 0$$

However, since for all *n* it holds that $P_n^Q(p(n)) = P_n^Q(p_n^Q(\alpha)) = \alpha$, $0 < \alpha < 1$, the former case violates condition (*) and the latter case condition (**) in the definition of a threshold function.

" \Leftarrow " Assume then that t(n) is *not* a threshold function for Q. Then there are either a sequence n_1, n_2, \ldots and a constant $\alpha < 1$ such that

$$p(n_k)/t(n_k) \to \infty$$
 but $P^Q_{n_k}(p(n_k)) \le \alpha$,

or a sequence n_1, n_2, \ldots and a constant $\alpha > 0$ such that

$$p(n_k)/t(n_k) \rightarrow 0$$
 but $P_{n_k}^Q(p(n_k)) \ge \alpha$.

In the former case,

$$t(n_k) \prec p(n_k) \leq p_{n_k}^Q(\alpha),$$

and in the latter case

$$t(n_k) \succ p(n_k) \ge p_{n_k}^Q(\alpha).$$

Thus in either case, $t(n) \not\approx p_n^q(\alpha)$ for some $0 < \alpha < 1$. \Box

Theorem 7.9 Every monotone graph property Q has a threshold function.

Proof. For brevity, denote $p_n^Q(\alpha) = p(\alpha)$. Choose some arbitrary $0 < \alpha < \frac{1}{2}$. The goal is to prove that $p(\alpha) \approx p(1-\alpha)$, thus establishing e.g.

$$t(n) = p\left(\frac{1}{2}\right) = p_n^Q\left(\frac{1}{2}\right)$$

as a threshold function for Q. (Since $p(\alpha) \le p(\frac{1}{2}) \le p(1-\alpha)$.)

Let $m \in \mathbb{N}$ be such that $(1 - \alpha)^m \leq \alpha$. Let $p = p_n(\alpha)$ and consider a sample of *m* independent graphs G_1, \ldots, G_m from $\mathcal{G}(n, p)$. Then the graph $G_1 \cup \cdots \cup G_m \in \mathcal{G}(n,q)$, where $q = 1 - (1-p)^m \leq mp$, and so

$$\Pr(G_1 \cup \cdots \cup G_m \text{ has } Q) \leq \Pr(G \in \mathcal{G}(n, mp_n(\alpha)) \text{ has } Q).$$

On the other hand, since Q is monotone, if any G_i has Q, then so does $G_1 \cup \cdots \cup G_m$. Thus,

$$\Pr(G_1 \cup \cdots \cup G_m \text{ does not have } Q) \le (1 - \Pr(G_i \text{ has } Q))^m = (1 - \alpha)^m \le \alpha.$$

7. Random Graphs

Hence,

$$\Pr_n^Q(mp_n(\alpha)) \ge \Pr(G_1 \cup \cdots \cup G_m \text{ has } Q) \ge 1 - \alpha,$$

and so

$$p_n(\alpha) \leq p_n(1-\alpha) \leq mp_n(\alpha),$$

i.e. $p(\alpha) \approx p(1-\alpha)$. (Since *m* depends only on α , not on *n*.)

Consider a graph property *Q* defined as "*G* has *Q*" if X(G) > 0, where $X \ge 0$ is a random variable on $\mathcal{G}(n, p)$.

E.g. if X(G) denotes the number of spanning trees of G, then property Q corresponds to connectedness.

Recall the two properties characterising a threshold function t(n):

- (i) $p(n) \prec t(n) \Rightarrow \text{almost no } G \in \mathcal{G}(n, p(n)) \text{ has } Q.$
- (ii) $p(n) \succ t(n) \Rightarrow \text{almost all } G \in \mathcal{G}(n, p(n)) \text{ have } Q.$

If X is integral, then one can aim to verify conditions (i) and (ii) by the so called "first-moment method" and "second-moment method", respectively.

The first-moment method consists simply of upper-bounding the expectation E[X] and applying Markov's inequality:

$$\Pr(X \ge 1) \le E[X] \qquad (\text{ more generally, for } a > 0$$
$$p(X \ge a) \le E[X]/a).$$

More specifically, one aims to show that if the choice of edge probabilities satisfies $p(n) \prec t(n)$, then $E[X_n] \to 0$. By Markov's inequality it then follows that also $P_n^Q(p(n)) = \Pr(X_n \ge 1) \to 0$.

The second-moment method is based on lower-bounding E[X] and upper-bounding Var[X].

Denote $\mu_n = E[X_n]$, $\sigma_n^2 = \text{Var}[X_n] = E[(X_n - \mu_n)^2] = E[X_n^2] - \mu_n^2$. Recall Chebyshev's inequality (a simple consequence of Markov's inequality): for any $\lambda > 0$,

$$\Pr(|X-\mu| \ge \lambda) \le \frac{\sigma^2}{\lambda^2}.$$

Lemma 7.10 If $\mu_n > 0$ for n large, and $\frac{\sigma_n^2}{\mu_n^2} \to 0$ as $n \to \infty$, then $\Pr(X_n > 0) \to 1$ as $n \to \infty$.

Proof. If $X_n = 0$, then $|X_n - \mu_n| = \mu_n$. Hence

$$\Pr(X_n=0) \le \Pr(|X_n-\mu_n| \ge \mu_n) \le \frac{\sigma_n^2}{\mu_n^2} \to 0 \text{ as } n \to \infty. \quad \Box$$

For the next result, denote the number of nodes in a graph G by |G|, the number of edges by e(G), and define its *density* as $\delta(G) = \frac{e(G)}{|G|}$. Any that a graph G is *balanced* if $\delta(G') \leq \delta(G)$ for all subgraphs G' of G.

Theorem 7.11 Let H be a balanced graph. Then the graph property "G has a subgraph isomorphic to H" has threshold function $n^{-1/\delta(H)}$.

Proof. Denote X(G) =number of H-subgraphs of a given graph G. Let k = |H|, l = e(H), so $\delta(H) = l/k$, and let $G \in \mathcal{G}(n, p)$, where $p = \gamma n^{-1/\delta(H)} = \gamma n^{-k/l}$ for some $\gamma = \gamma_n$. Let us first apply the first-moment method to show that if $\gamma \to 0$, then almost no G contains a subgraph isomorphic to H. Denote

 $\mathcal{H} = \{ \text{all copies of } H \text{ on vertex-set of } G \}.$

Then $|\mathcal{H}| = \binom{n}{k}h \le \binom{n}{k}k! \le n^k$, where *h* is the number of different arrangements of *H* on a set of *k* vertices, $h = k!/|\operatorname{Aut}(H)|$. Thus

$$egin{aligned} E[X] &= \sum_{H' \in \mathcal{H}} \Pr(H' \subseteq G) = |\mathcal{H}| \cdot p^l \ &\leq n^k p^l = n^k (\gamma n^{-k/l})^l = \gamma^l \xrightarrow[\gamma
ightarrow 0, \end{aligned}$$

and by Markov's inequality the desired result follows.

For the other part, we wish apply the second-moment method to show that if $\gamma \rightarrow \infty$, then almost every graph *G* contains a subgraph isomorphic to *H*. For this, we need to verify that $\mu = E[X] > 0$ for all sufficiently large *n*, and then show that

$$\frac{\sigma^2}{\mu^2} = \frac{1}{\mu^2} (E[X^2] - \mu^2) \to 0 \quad \text{as } n \to \infty.$$

The first condition is easy to check: without loss of generality, assume that $\gamma = \gamma_n \ge 1$ for all *n*. Then:

$$\mu = E[X] = |\mathcal{H}| \cdot p^{l}$$
$$= {n \choose k} h \cdot \gamma_{n}^{l} \cdot n^{-k}$$
$$\geq \operatorname{const} \cdot n^{k} \cdot h \cdot \gamma_{n}^{l} \cdot n^{-k}$$
$$> 0.$$

7. Random Graphs

For the other requirement, let us try to compute:

$$\begin{split} E[X^2] &= \sum_{H',H''\in\mathcal{H}} \Pr(H'\cup H''\subseteq G) \\ &= \sum_{H',H''\in\mathcal{H}} p^{e(H')+e(H'')-e(H'\cap H'')} \\ &\leq \sum_{H',H''\in\mathcal{H}} p^{2l-i\delta(H)}, \end{split}$$

where $i = |H' \cap H''|$. (Note that $\delta(H' \cap H'') \le \delta(H)$.) Denote then $\mathcal{H}_i^2 = \{(H', H'') \in \mathcal{H}^2 : |H' \cap H''| = i\}$ and compute separately for each *i* the sum

$$A_i = \sum_{{}_{\mathcal{H}_i}^2} \Pr(H' \cup H'' \subseteq G)$$

Case i = 0:

$$A_{0} = \sum_{\mathcal{H}_{0}^{2}} \Pr(H' \cup H'' \subseteq G)$$

= $\sum_{\mathcal{H}_{0}^{2}} \Pr(H' \subseteq G) \cdot \Pr(H'' \subseteq G)$ $H', H'' \text{ independent}$
 $\leq \sum_{\mathcal{H}^{2}} \Pr(H' \subseteq G) \cdot \Pr(H'' \subseteq G)$
= $\left(\sum_{\mathcal{H}} \Pr(H' \subseteq G)\right)^{2}$
= μ^{2} .

Case
$$i \ge 1$$
.

$$A_{i} = \sum_{\mathcal{H}_{i}^{2}} \Pr(H' \cup H'' \subseteq G)$$

$$= \sum_{H' \in \mathcal{H}} \sum_{\substack{H'': \\ |H' \cap H''| = i}} \Pr(H' \cup H'' \subseteq G)$$

$$\leq |\mathcal{H}| \cdot \binom{k}{i} \binom{n-k}{k-i} hp^{2l} p^{-il/k} \qquad h = \frac{k!}{|\operatorname{Aut}(H)|}$$

$$\leq |\mathcal{H}| \cdot c_{1} n^{k-i} hp^{2l} (\gamma n^{-k/l})^{-il/k}$$

$$= \mu \cdot c_{1} n^{k-i} hp^{l} \gamma^{-il/k} n^{i}$$

$$= \mu c_{2} \binom{n}{k} hp^{l} \gamma^{-il/k}$$

$$= \mu c_{2} \binom{n}{k} hp^{l} \gamma^{-il/k}$$

$$\leq \mu^{2} \cdot c_{2} \gamma^{-il/k}.$$

Thus, denoting $c_3 = kc_2$, we get the estimate

$$\frac{E[X^2]}{\mu^2} = \left(\frac{A_0}{\mu^2} + \frac{\sum_i A_i}{\mu^2}\right) \le 1 + c_3 \gamma^{-l/k}$$

and hence

$$\frac{\sigma^2}{\mu^2} = \frac{E[X^2] - \mu^2}{\mu^2} \le c_3 \gamma^{-l/k} \xrightarrow[\gamma \to \infty]{} 0.$$

The desired result then follows by Lemma 7.10. \square

Corollary 7.12 For $k \ge 3$, the property of containing a k-cycle has threshold $t(n) = n^{-1}$. (Note that the threshold is independent of k.)

Corollary 7.13 For $k \ge 2$, the property of containing a specific tree structure T on k nodes has threshold function $t(n) = n^{-k/(k-1)}$.

Corollary 7.14 For $k \ge 2$, the property of containing a k-clique ($\approx K_k$) has threshold function $t(n) = n^{-2/(k-1)}$. \Box

Denote $\delta^*(H) = \max{\{\delta(H') | H' \text{ is subgraph of } H\}}$.

Theorem 7.11' The graph property "G has a subgraph isomorphic to H" has threshold function $n^{-1/\delta^*(H)}$.

80